

#### LA-UR-21-24096

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Title: LUNA Condition-Based Monitoring Update: Auto-encoders on Actuator Data

Author(s): Green, Andre Walter

Intended for: Progress report to sponsor.

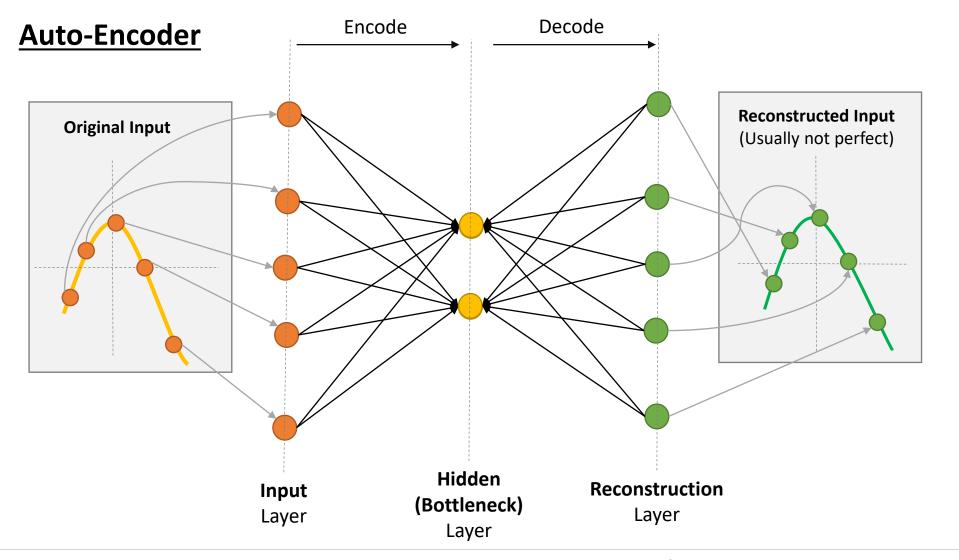
Issued: 2021-04-28



# **LUNA Condition-Based Monitoring Update:**

# Auto-encoders on Actuator Data

Presented 4/20/2021

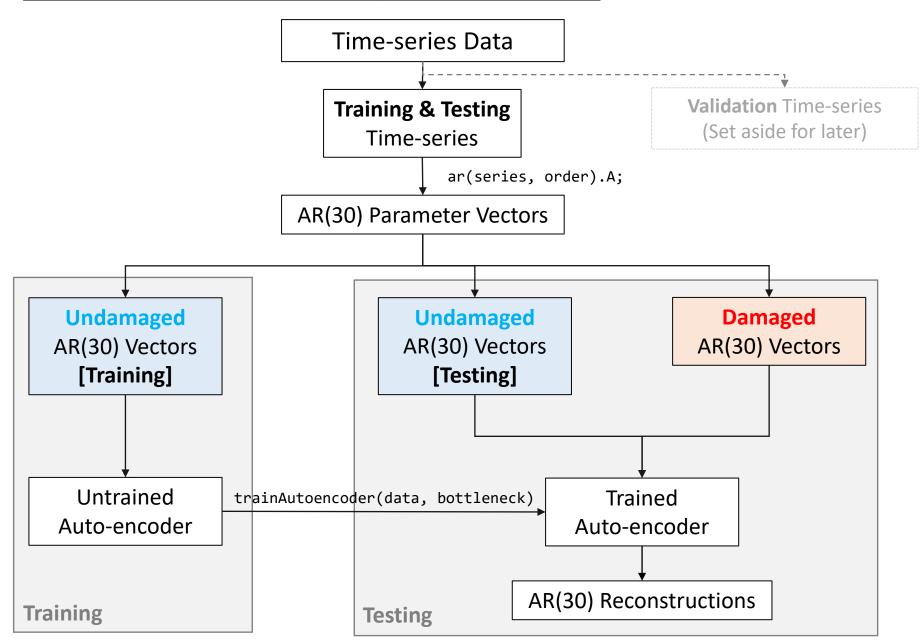


An auto-encoder is a neural network trained to output a reconstruction of its input.

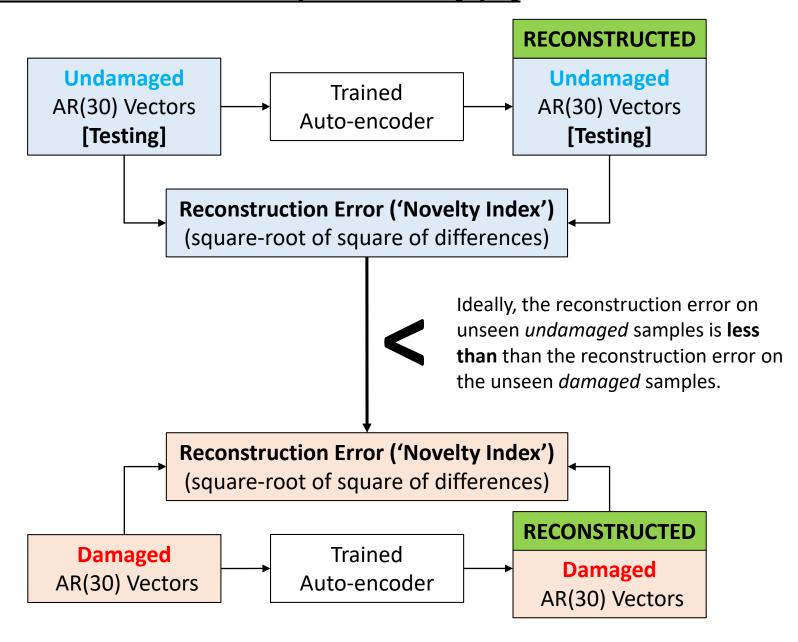
There is a 'bottleneck' layer whose dimension is typically lower than that of the input.

Because the input passes through the bottleneck, the auto-encoder must learn to compress (encode) the input into this low-dimensional form, and to decompress (decode) the compressed data into a close approximation of its original form.

#### **Auto-encoders for Anomaly Detection [1/2]**



#### **Auto-encoders for Anomaly Detection [2/2]**

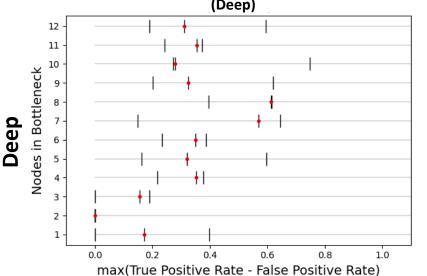


#### **Bottleneck Optimization Performance Summary**

#### **All Actuators**

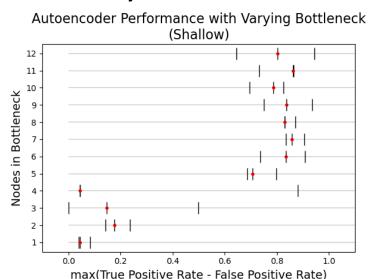
Autoencoder Performance with Varying Bottleneck (Shallow) 12 11 10 Nodes in Bottleneck 9 8 5 3 2 0.4 0.6 0.8 1.0 0.0 max(True Positive Rate - False Positive Rate)

Autoencoder Performance with Varying Bottleneck (Deep)

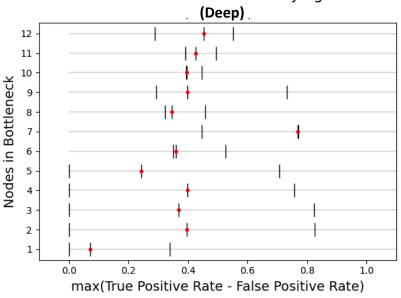


**True Positive:** The reconstruction error for a sample is above a threshold and the sample is damaged.

#### **Only Board 401 Data**



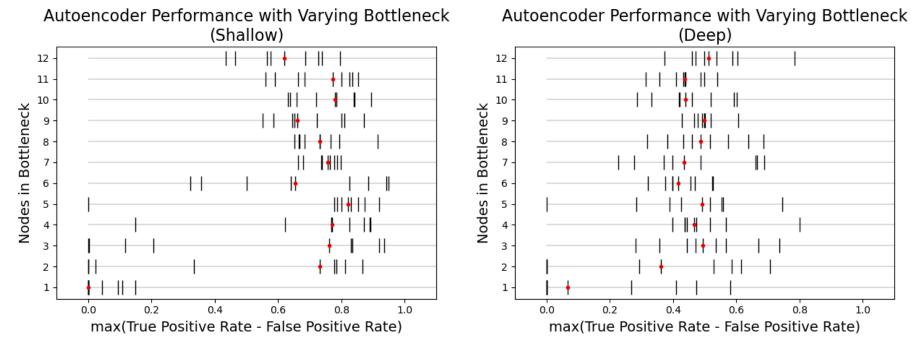
Autoencoder Performance with Varying Bottleneck



These graphs use 3-fold validation.

#### **Bottleneck Optimization Performance Summary**

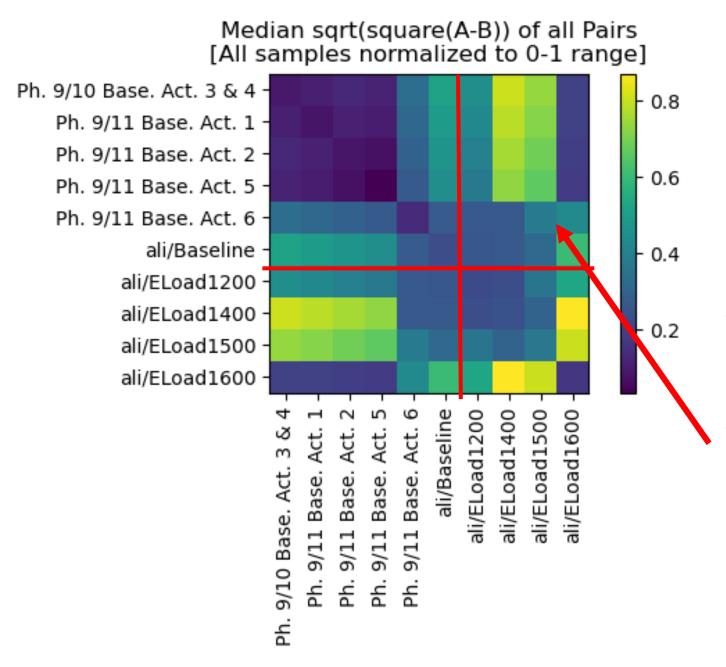
#### **Just Board 401 Data**



Both deep and shallow appear to do better when not using the Philadelphia data to supplement the baseline data under the current method of normalization.

So far the shallow networks appears to perform slightly better.

#### Between-Class and Within-Class Median sqrt(squared(differences)))



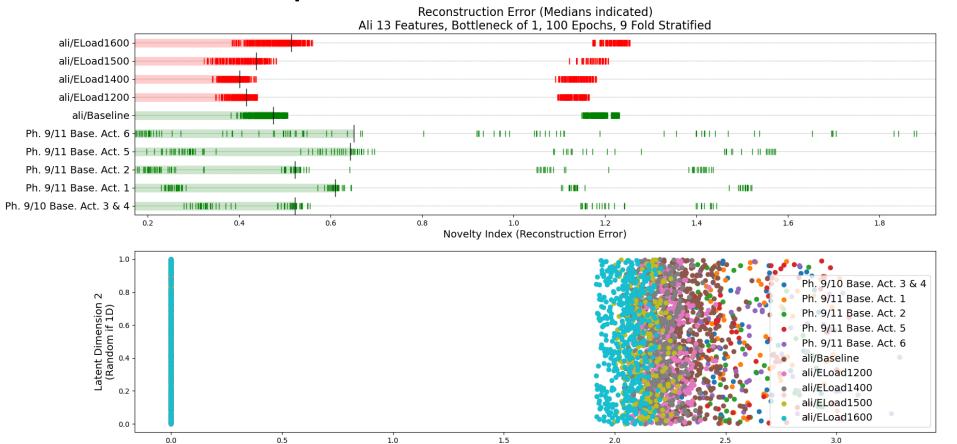
If the reconstruction error is above ~0.2, then the error in reconstruction is greater than difference between some damaged and undamaged classes.

ELoad1600 and the Philadelphia datasets are surprisingly similar.

This may be due to the normalization scheme used (global min/max, normalized to 0-1 range).

The top-right quadrant (mirrored by bottom left) shows the median differences between undamaged and damaged classes.

# **Multi-Actuator Shallow Auto-encoder**



Latent Dimension 1

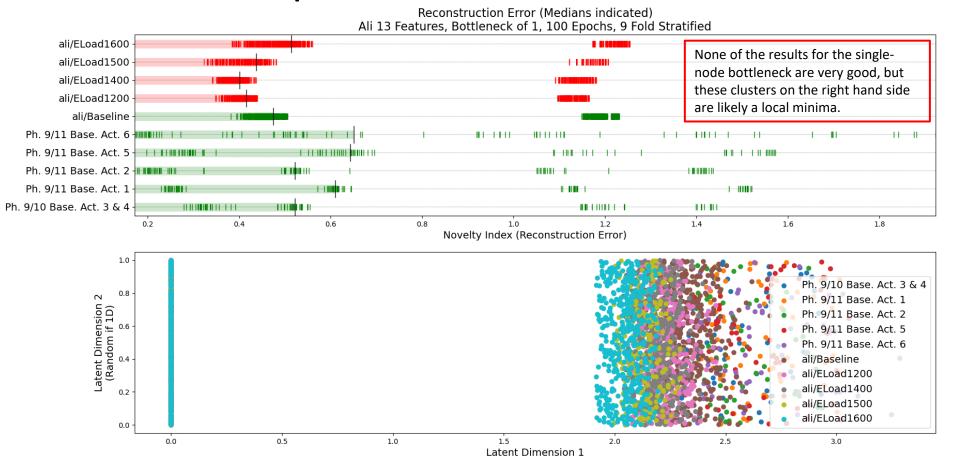
#### All of the following graphs use 9-fold\* stratified cross-validation.

Each consists both of a separate training & testing set and an auto-encoder. Each fold has approximately 1039 training samples and 377 testing samples.

Because the 9 auto-encoders are trained separately, it's possible for them to learn different methods for reconstructing the data. Sometimes the auto-encoder will get stuck in local minima, and perform poorly.

It's also possible the auto-encoder is learning a good representation, but one that doesn't produce different magnitude reconstruction errors for damaged/undamaged inputs.

<sup>\*9-</sup>Fold was selected because 1/10<sup>th</sup> of the data has been set aside for eventual validation.



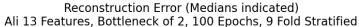
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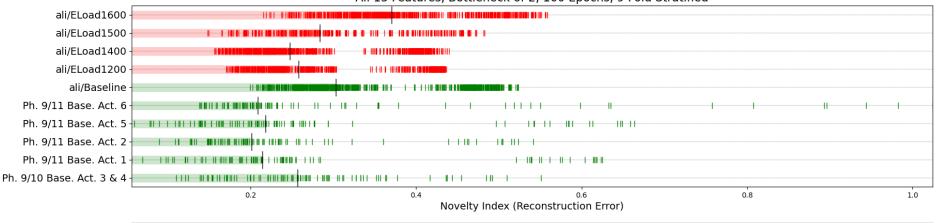
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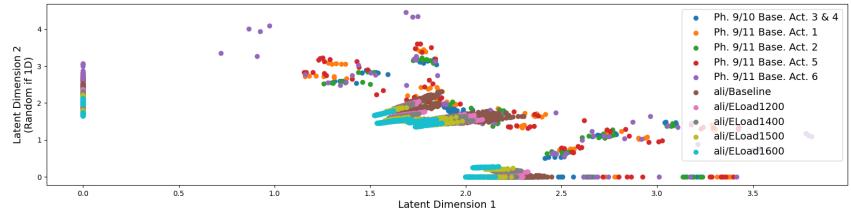
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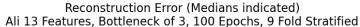
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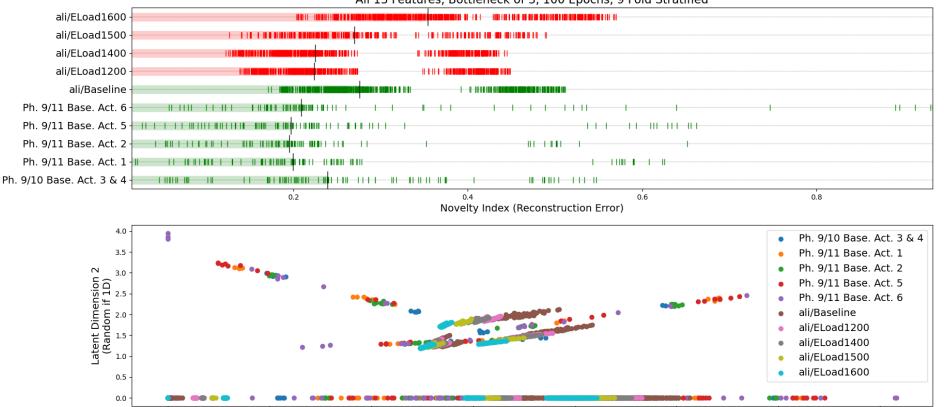
<sup>\*9-</sup>Fold was selected because 1/10<sup>th</sup> of the data has been set aside for eventual validation.











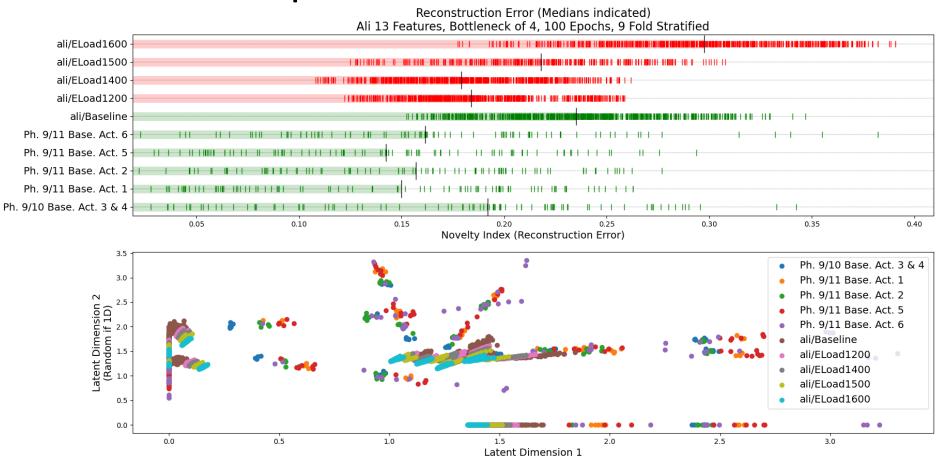
1.5

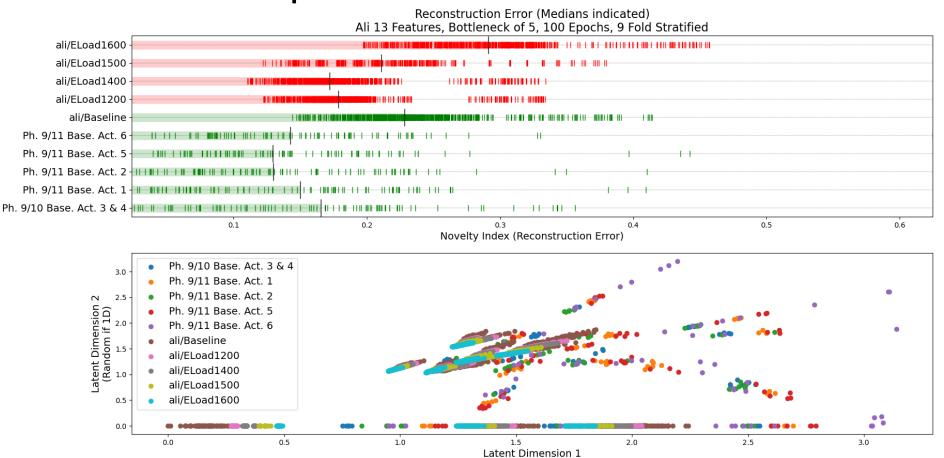
2.0

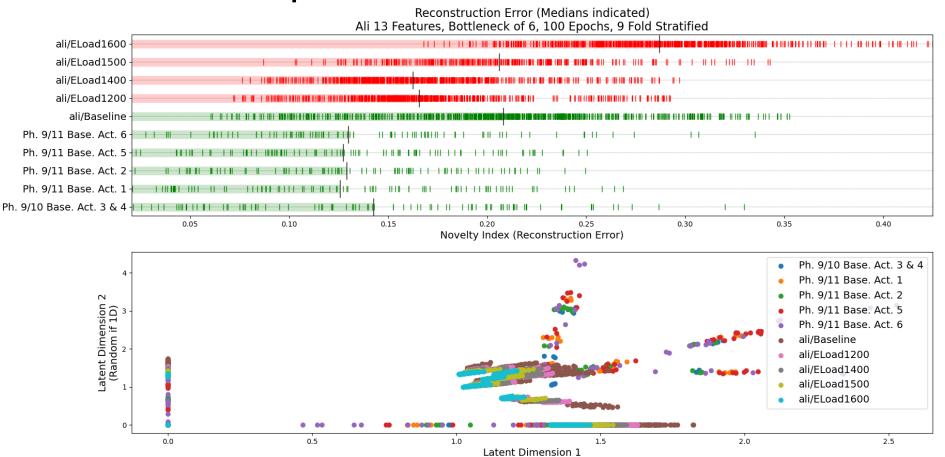
Latent Dimension 1

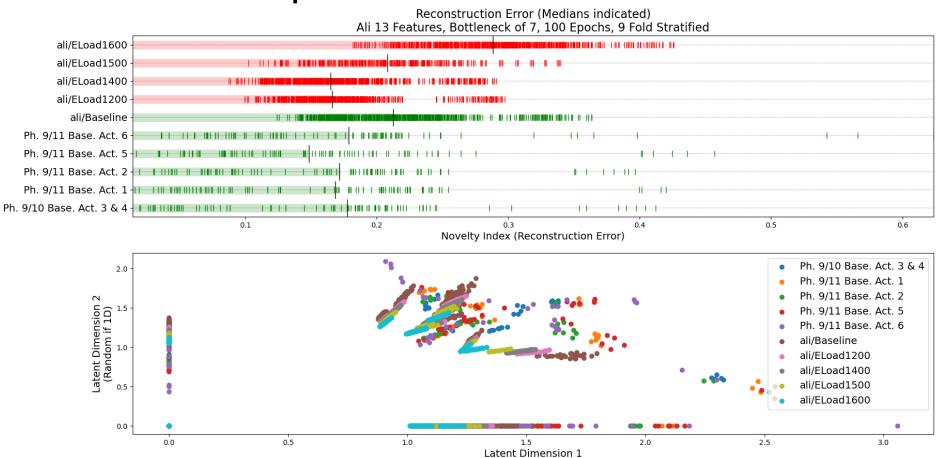
2.5

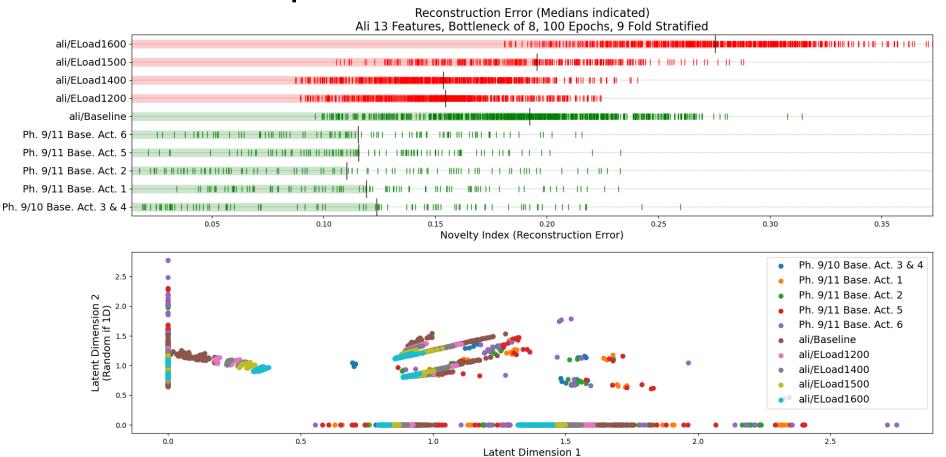
3.0

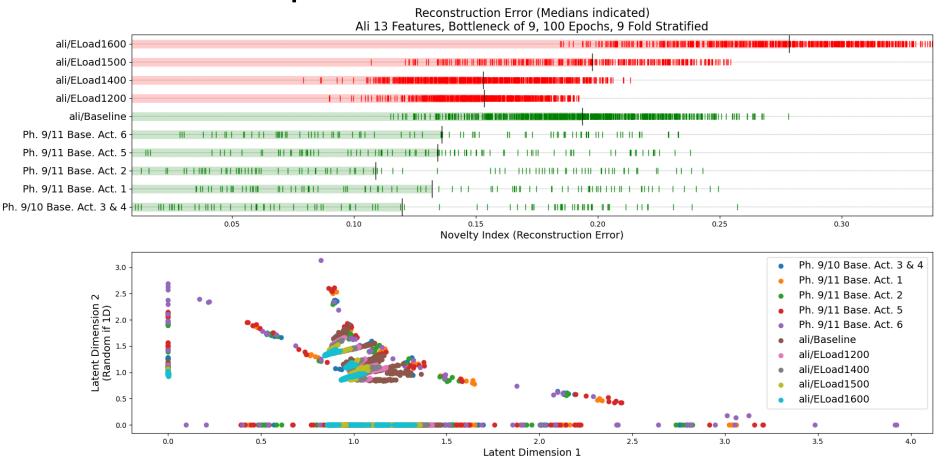


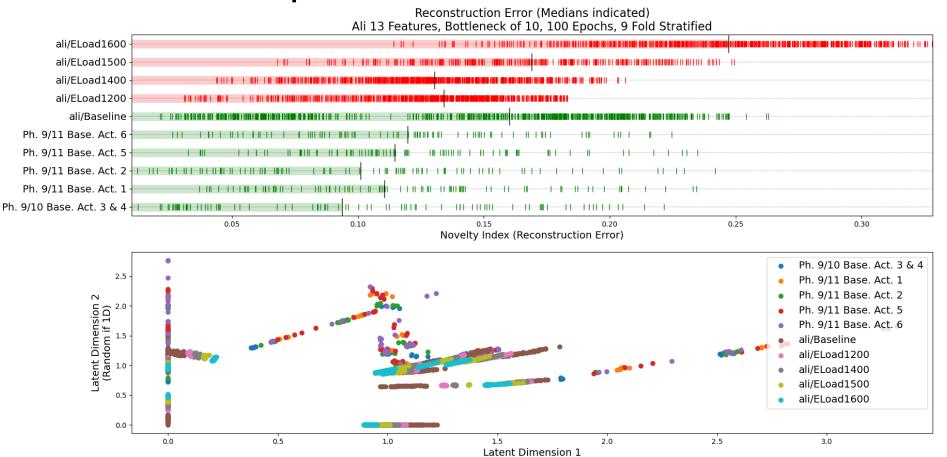


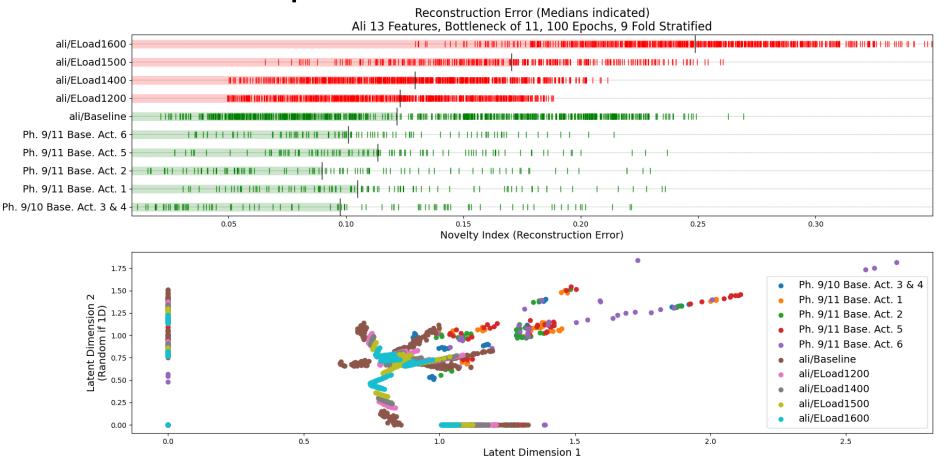


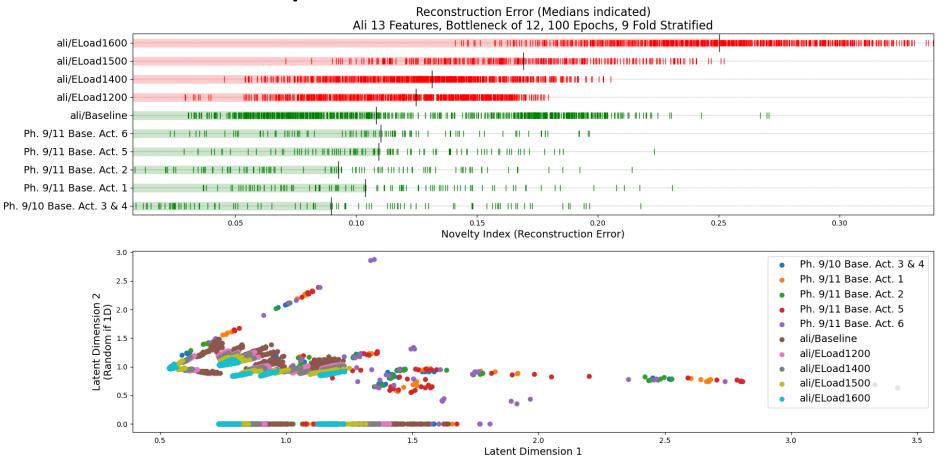




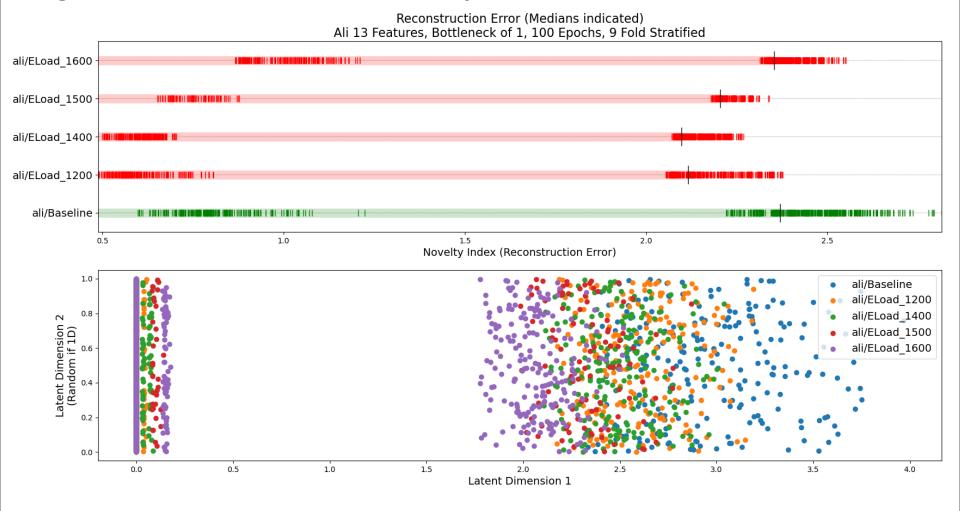


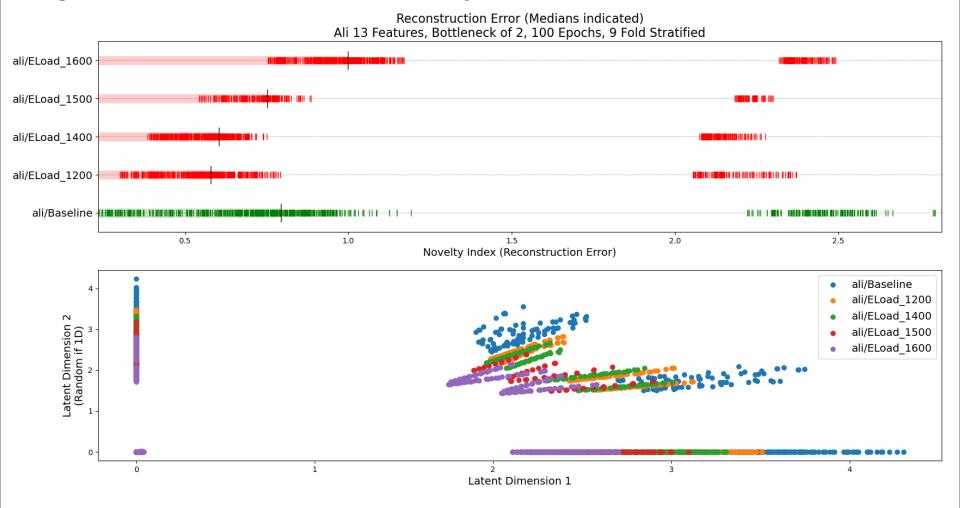


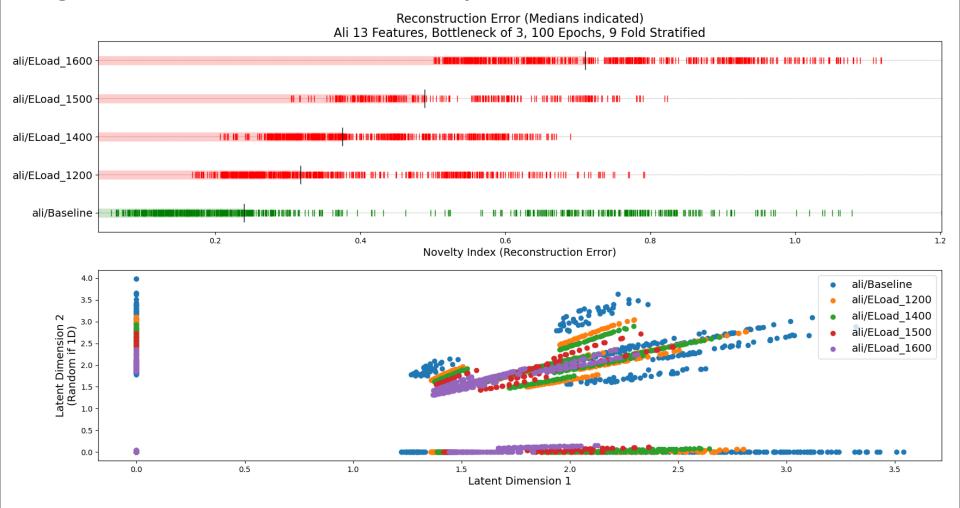


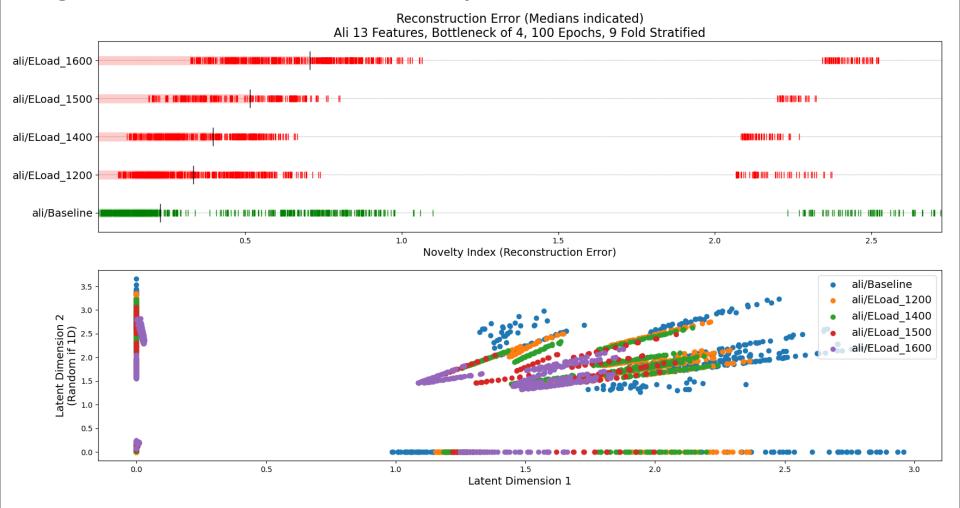


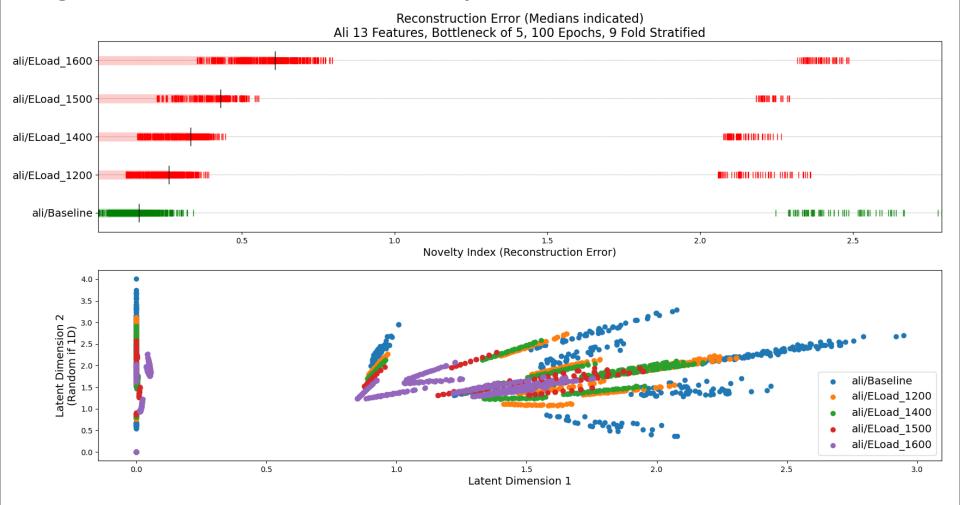
**Single-Actuator Shallow Auto-encoder** 

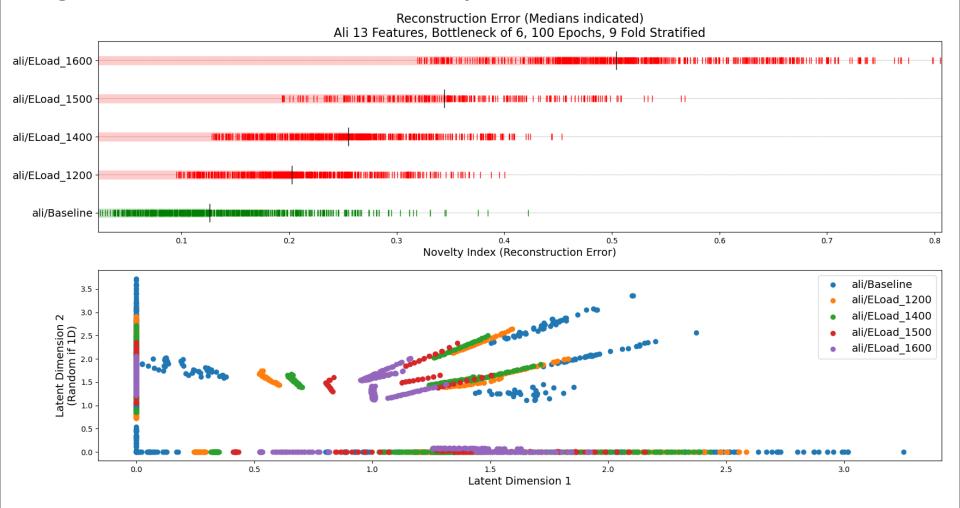


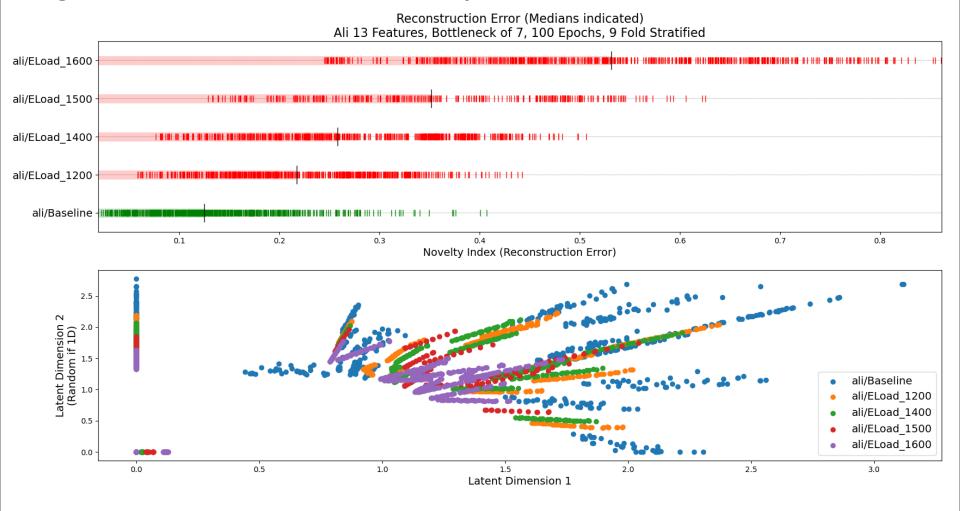


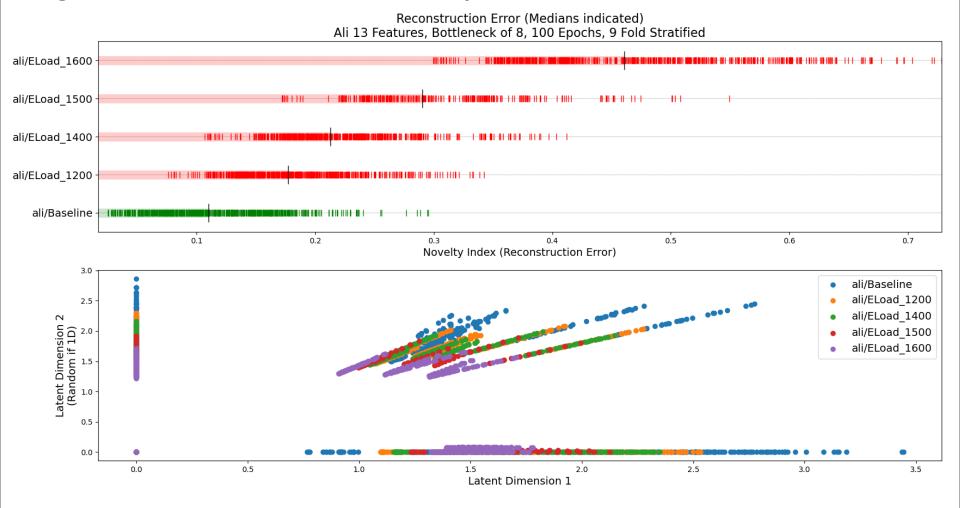


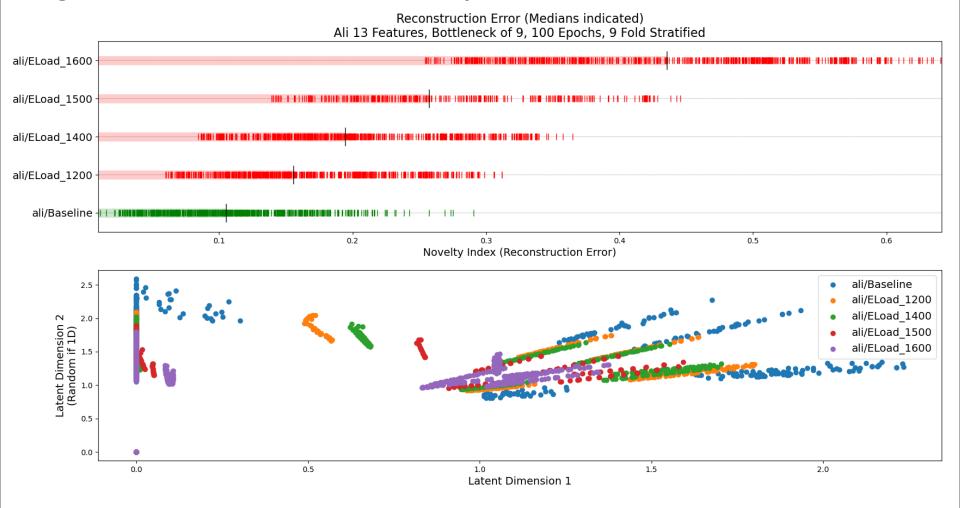


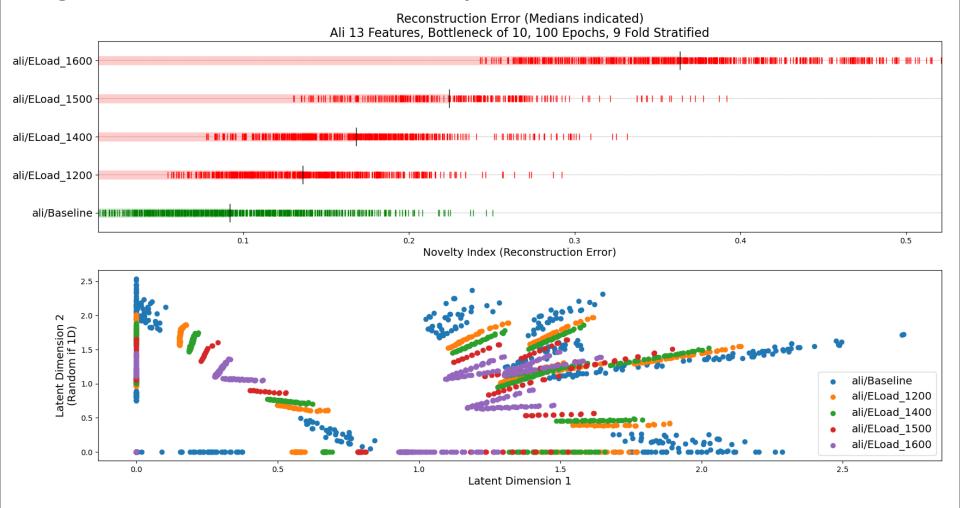


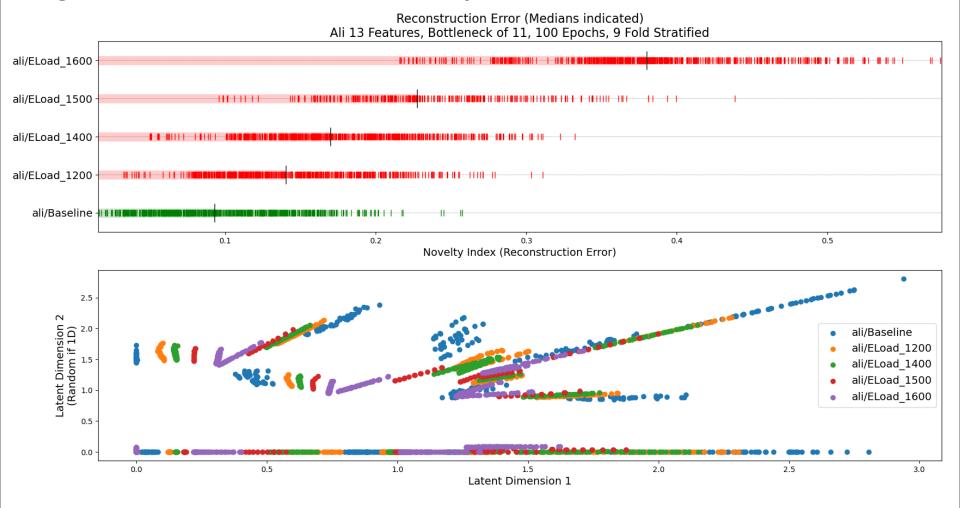


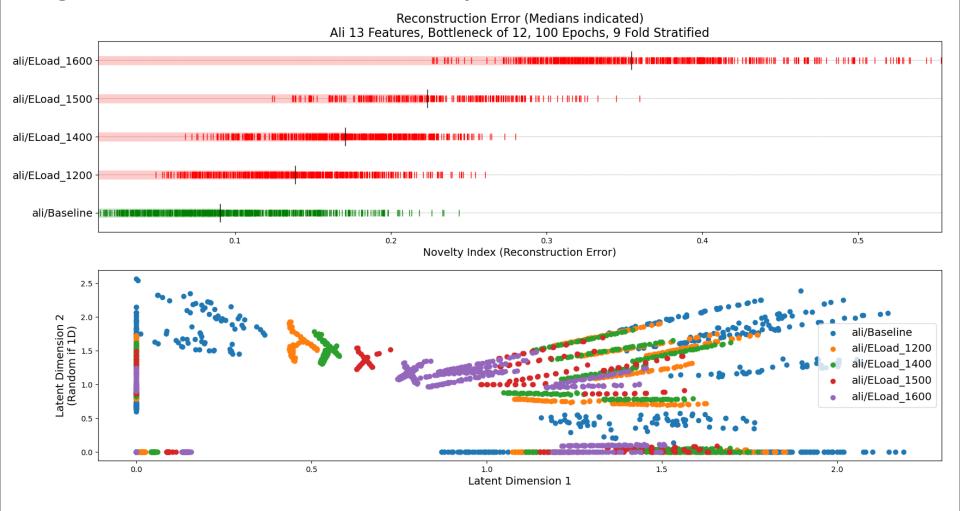




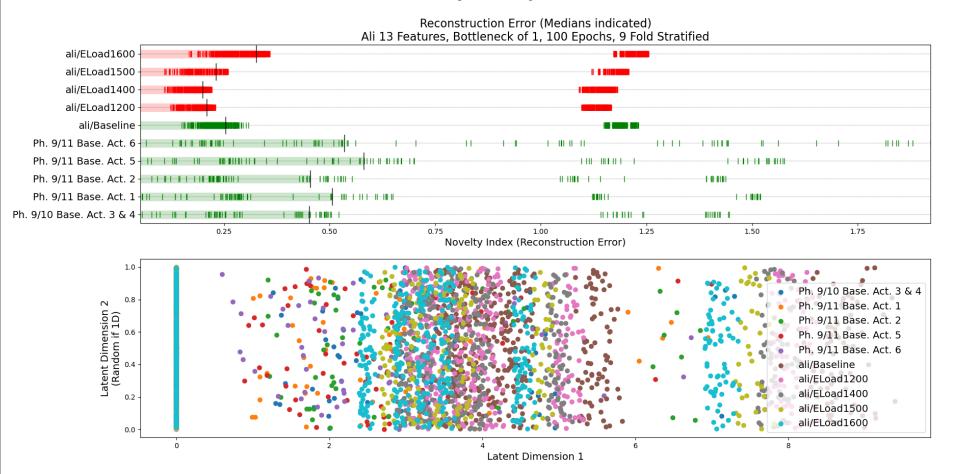


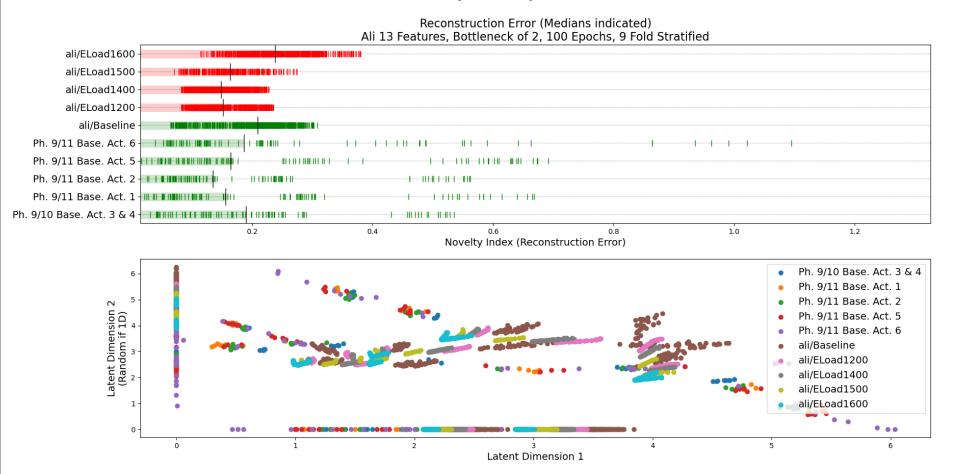


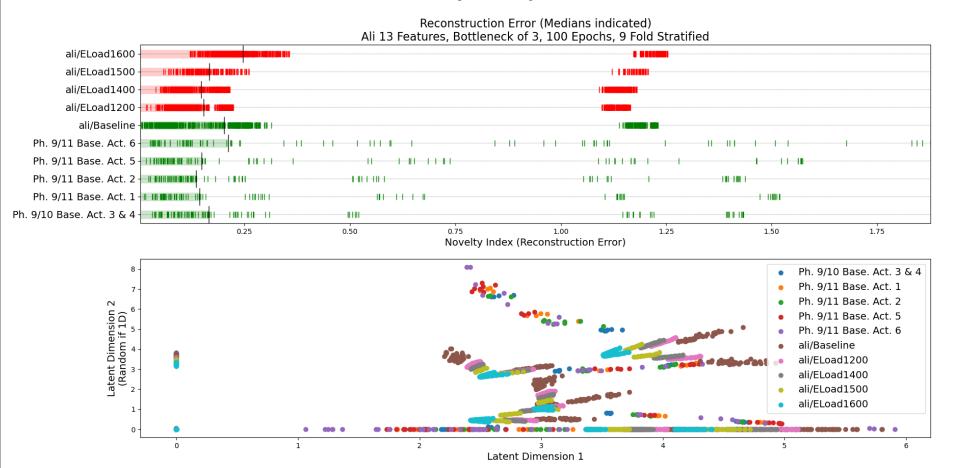


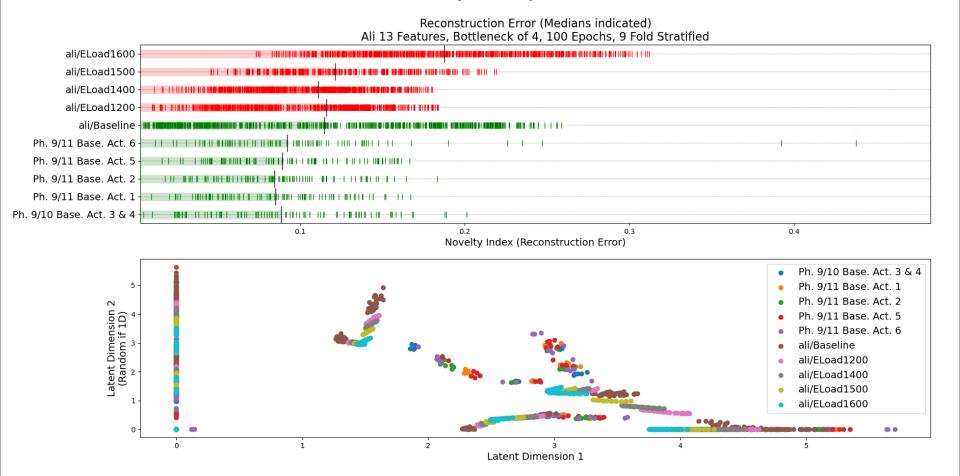


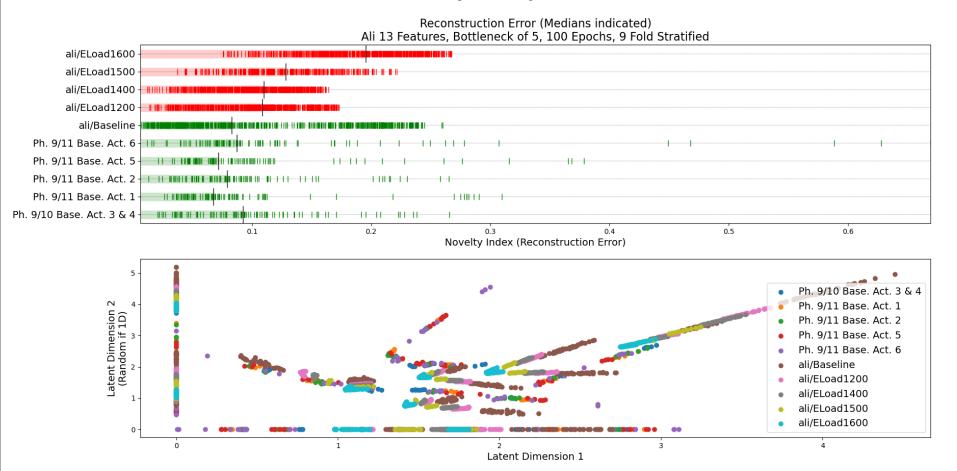
**Multi-Actuator Deep Auto-encoder** 

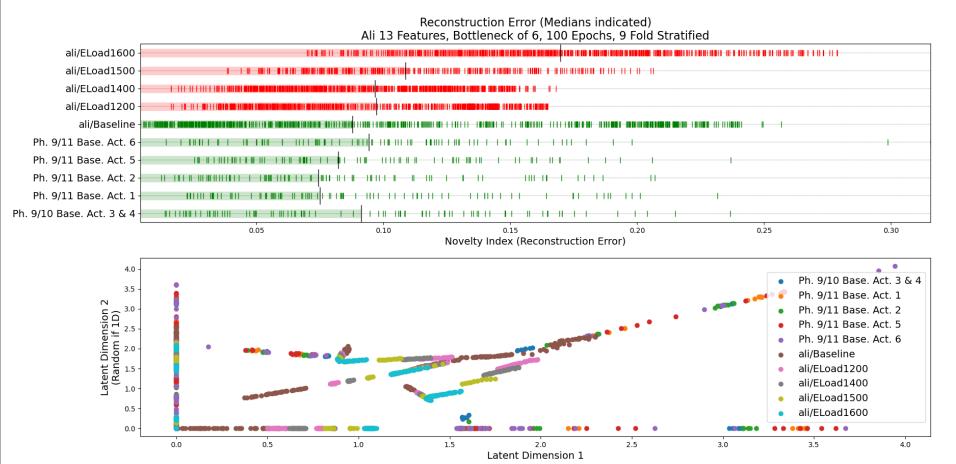


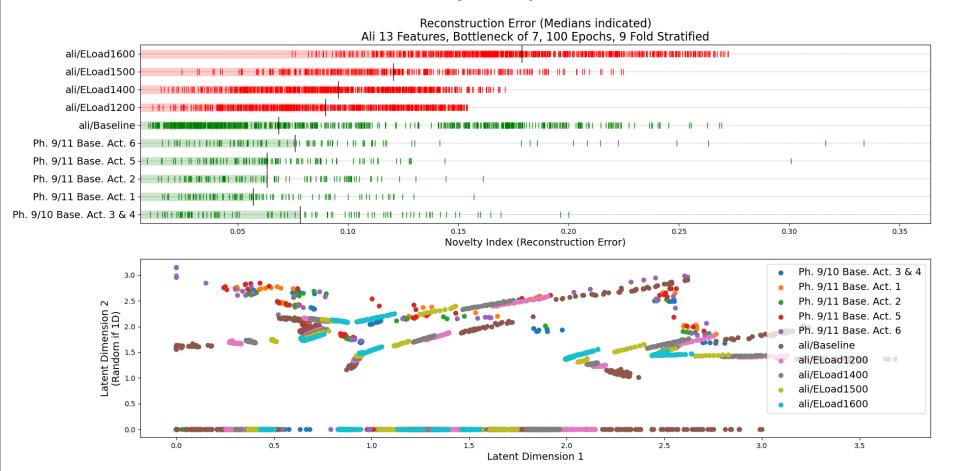


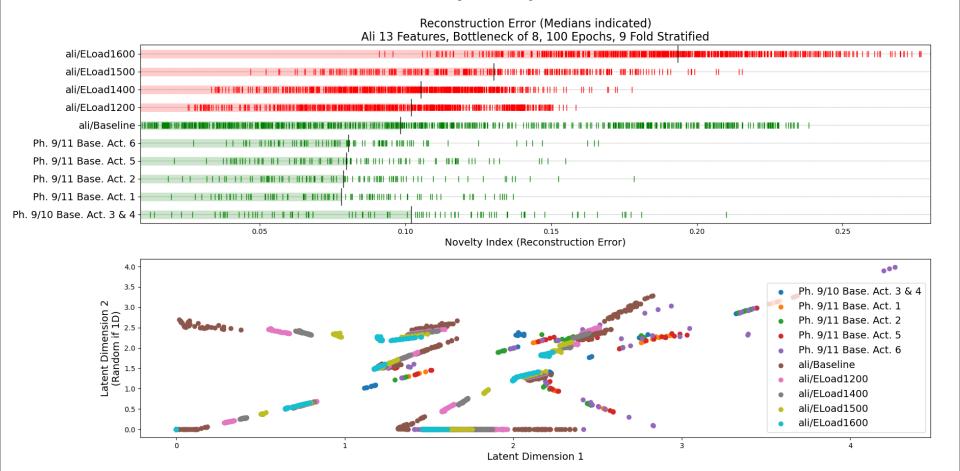


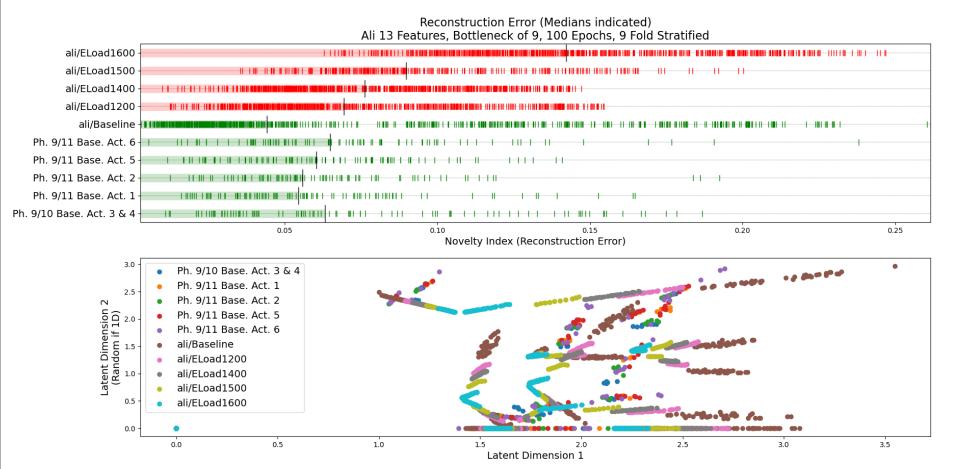


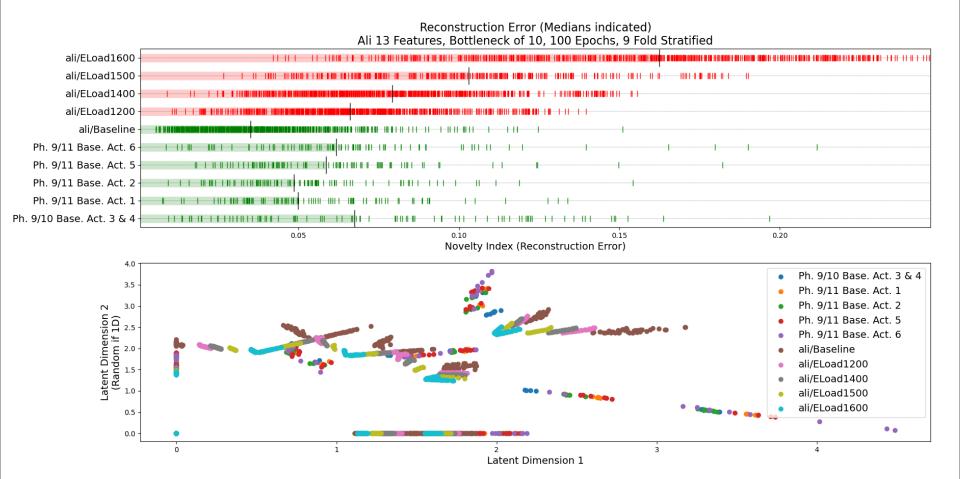


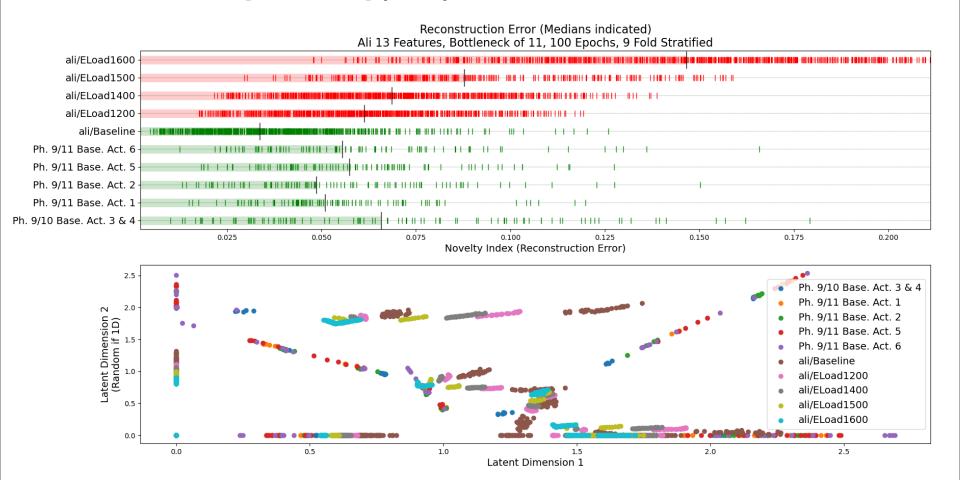


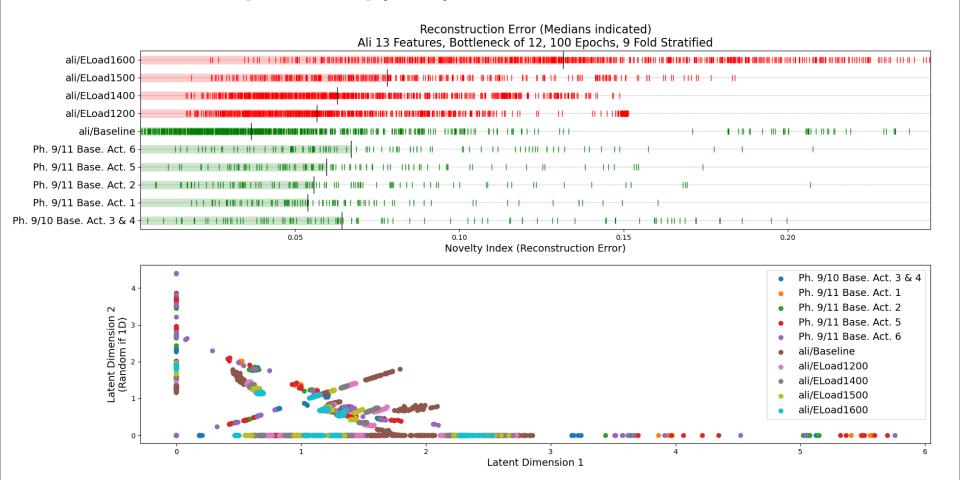












**Single-Actuator Deep Auto-encoder** 

